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# Quantum algorithms for linear systems of equations

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based on joint work  
with Rolando Somma

# Outline

- Motivation
- HHL and CKS Algorithms
- New Adiabatic like algorithms
- Summary



# Motivation

- Quantum computers have the potential to perform computations that are classically intractable.
- Fast quantum algorithms exist for simulating the dynamics of quantum systems and factoring integers.
- Some problems cannot be solved dramatically faster by quantum computers than by classical ones.
- The full power of quantum computers is unknown.
- It is important to find fast quantum algorithms for problems with broad applications.



## Problem: System of linear equations

Given an  $N \times N$  matrix  $A$ , an  $N \times 1$  vector  $\vec{b}$  and the equation

$$\begin{bmatrix} A_{11} & A_{12} & \dots \\ \vdots & \ddots & \\ A_{N1} & & A_{NN} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_N \end{bmatrix} = \begin{bmatrix} b_1 \\ \vdots \\ b_N \end{bmatrix}$$

that is  $A\vec{x} = \vec{b}$ , solve for  $\vec{x}$ .

This type of problem is ubiquitous in scientific computing and engineering applications.

Classical methods require at least  $O(N)$  operations. ( $A^{-1}\vec{b} = \vec{x}$ )



# Harrow Hassidim Lloyd (HHL) Algorithm

*Phys. Rev. Lett. 103, 150502 (2009)*

Convert vectors to states living in Hilbert space  $n = \log_2(N)$  qubits.

$$\vec{b} \rightarrow |b\rangle = \frac{\sum_i b_i |i\rangle}{\sum_i |b_i|^2}, \quad \vec{x} \rightarrow |x\rangle = \frac{\sum_i x_i |i\rangle}{\sum_i |x_i|^2}$$

Assume  $A$  is Hermitian. If not use instead

$$\tilde{A} = \begin{bmatrix} 0 & A \\ A^\dagger & 0 \end{bmatrix} = \tilde{A}^\dagger$$

Then the solution is *given by*

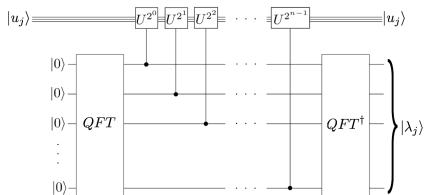
$$|x\rangle = \frac{A^{-1} |b\rangle}{\|A^{-1} |b\rangle\|}$$



Let us write  $|b\rangle$  in the eigenbasis of  $A$

$$|b\rangle = \sum \beta_j |u_j\rangle, \quad A|u_j\rangle = \lambda_j |u_j\rangle$$

$$A^{-1} |b\rangle = \sum \beta_j \lambda_j^{-1} |u_j\rangle$$



Phase estimation algorithm

Hamiltonian simulation with  
 $U = e^{iAt_0/T}$ .

$$|b\rangle |0\rangle_{\text{anc}} \rightarrow \sum \beta_j |u_j\rangle |\lambda_j\rangle_{\text{anc}}$$

Next we would like to perform the linear map

$$|\lambda_j\rangle_{\text{anc}} \rightarrow C \lambda_j^{-1} |\lambda_j\rangle_{\text{anc}}$$





$|\lambda_j\rangle_{\text{anc}} \rightarrow C\lambda_j^{-1} |\lambda_j\rangle_{\text{anc}}$  is not unitary but can be implemented with finite probability

$$\sum \beta_j |u_j\rangle |\lambda_j\rangle_{\text{anc}} |0\rangle_{\text{a}} \rightarrow \sum \beta_j |u_j\rangle |\lambda_j\rangle_{\text{anc}} \left( \frac{C}{\lambda_j} |1\rangle_{\text{a}} + \sqrt{1 - \frac{C^2}{\lambda_j^2}} |0\rangle_{\text{a}} \right)$$

If the ancillary qubit is measured and found to be in state 1, then we get the state

$$\propto \sum \frac{\beta_j}{\lambda_j} |u_j\rangle |\lambda_j\rangle_{\text{anc}}$$

Finally we undo the phase estimation step to get:

$$|x\rangle \propto \sum \frac{\beta_j}{\lambda_j} |u_j\rangle$$



Let  $\kappa$  : condition number of  $A$   
 $d$  : number of nonzero entries per row  
 $\epsilon$  : precision with which to prepare  $|x\rangle$ .

The HHL algorithm takes  $\text{poly}(\log N, \kappa, 1/\epsilon, d)$  quantum steps to output  $|x\rangle$ , compared with  $\text{poly}(N, \kappa, \log(1/\epsilon), d)$  steps required to find  $\vec{x}$  using the best known method on a classical computer.

### Caveats:

- Finding full answer  $\vec{x}$  requires  $O(N)$  repetitions to measure the amplitudes of  $|x\rangle$ .
  - HHL can provide features of  $\vec{x}$  such as expectation values over sparse matrices  $\vec{x}^\dagger \cdot B \cdot \vec{x}$ .
- Input vector  $|b\rangle$  needs to be prepared (can dominate complexity)
- The matrix  $A$  must be well-conditioned:  $\kappa = \text{polylog}(N)$  and it must be efficient to simulate  $e^{iAt}$ .



# Linear Combination of Unitaries Approach

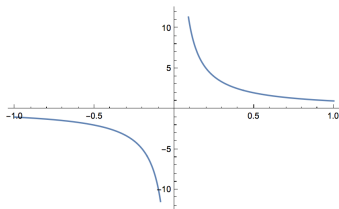
Childs, Kothari, Somma arXiv:1511.02306

Strategy: Express  $A^{-1}$  as a linear combination of easy-to-preform unitaries.

$$A^{-1} \approx \sum_t \alpha_t e^{-iAt}$$

Prepare ancillas in a state proportional to  $\sum_t \sqrt{\alpha_t} |t\rangle$ .

In order to create  $|x\rangle \propto A^{-1} |b\rangle$  using LCU requires the ability to implement  $C_U = \sum_t |t\rangle \langle t| \otimes e^{-iAt}$  and  $\text{Ref}(|b\rangle) = \mathbb{1} - 2 |b\rangle \langle b|$ .



The function blows up at the origin, but it is sufficient to approximate it in the domain  $[-1, \frac{1}{\kappa}] \cup [\frac{1}{\kappa}, 1]$

Bottomline: Runtime improved to  $\text{poly}(\log N, \log(1/\epsilon))$ .



## An adiabatic algorithm *(with Rolando Somma)*

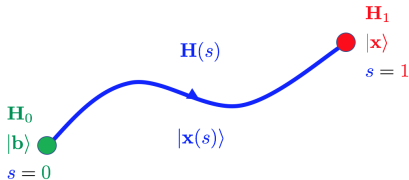
Consider the Hamiltonian

$$H := \frac{|b\rangle\langle b|}{\langle b|A^{-1}|b\rangle} + (\mathbb{1} - A),$$

$|x\rangle$  is unique eigenstate with largest eigenvalue 1

$$H(A^{-1}|b\rangle) = \frac{|b\rangle\langle b|A^{-1}|b\rangle}{\langle b|A^{-1}|b\rangle} + (\mathbb{1} - A)A^{-1}|b\rangle = A^{-1}|b\rangle$$

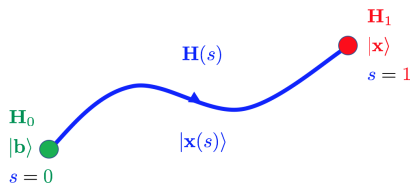
Then, the linear system of equations can be solved by an adiabatic evolution in which  $s$  is increased slowly in time or by randomization method.



We define an interpolation such that  $H(0) = |b\rangle \langle b|$ ,  $H(1) = H$ .

$$A(s) = (1 - s)\mathbb{1} + sA$$

$$H(s) = \frac{|b\rangle \langle b|}{\langle b| A(s)^{-1} |b\rangle} + (1 - A(s))$$



It is straightforward to show that

$$|x(s)\rangle = A(s)^{-1} |b\rangle / \|A(s)^{-1} |b\rangle\|$$

is the unique eigenstate of largest eigenvalue 1.



# Quantum Zeno Effect

By measuring  $H(s)$  for  $s_0 = 0 < s_1 < \dots < s_q = 1$ , it is possible to follow the adiabatic path to a given approximation  $\epsilon$  as long as:

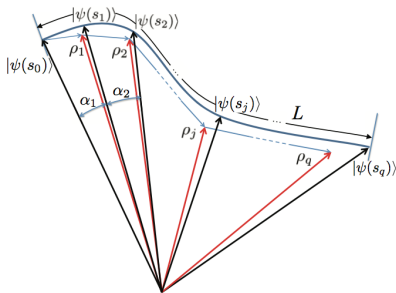
$$|\langle x(s_j) | x(s_{j+1}) \rangle|^2 < 1 - \epsilon/q$$

Need to space  $s_k$  such that change in  $|x(s)\rangle$  (eigenstate of  $H(s)$ ) from  $s_j$  to  $s_{j+1}$  is uniform and small.

We find that

$$s_j = \frac{1 - (1/\kappa)^{j/q}}{1 - 1/\kappa}$$

with  $q = \lceil \log^2(\kappa)/\epsilon \rceil$  is sufficient.



## Evolution Randomization Boixo, Knill, Somma (2009)

Evolving with  $H(s_k)$  for a random amount of time causes decoherence in the energy eigenbasis:

$$\rho = \sum_{\alpha, \beta} \rho_{\alpha\beta} |E_\alpha\rangle \langle E_\beta| \rightarrow \sum_{\alpha, \beta} \rho_{\alpha\beta} \int d\mu(t) e^{i(E_\beta - E_\alpha)t} |E_\alpha\rangle \langle E_\beta|$$

This effectively simulates a projective measurement if

$$\int d\mu(t) e^{i(E_\beta - E_\alpha)t} \approx \delta_{\alpha, \beta}$$

In fact we only need decoherence for one eigenstate  $|x(s)\rangle$  with eigenvalue 1.



$$H(s) = g(s) |b\rangle \langle b| + (\mathbb{1} - A(s)); \quad g(s) = \frac{1}{\langle b| A(s)^{-1} |b\rangle}$$

Assuming we know  $g(s)$ , sampling  $t_j$  uniformly according to

$$t_j \in [0, 1, 2, \dots, Q_j - 1], \quad Q_j = \lceil 2\pi / \Delta(s_j) \rceil$$

is sufficient.

The gap of  $H(s)$  satisfies  $\Delta(s) \geq 1 - s(1 - 1/\kappa)$ .

The total runtime of the algorithm  $T^* \leq \sum_{j=1}^q Q_j \leq 4\pi\kappa \log(\kappa)/\epsilon$ .



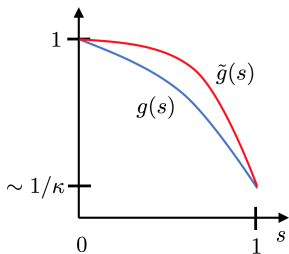


$$H(s) = g(s) |b\rangle \langle b| + (\mathbb{1} - A(s)); \quad g(s) = \frac{1}{\langle b| A(s)^{-1} |b\rangle}$$

The problem is that we don't know  $g(s)$ . It satisfies:

$$g(0) = 1; \quad g(1) = \langle b| A^{-1} |b\rangle^{-1}$$

First assume that we know  $g(1)$   
and find  $\tilde{g}(s)$  such that:



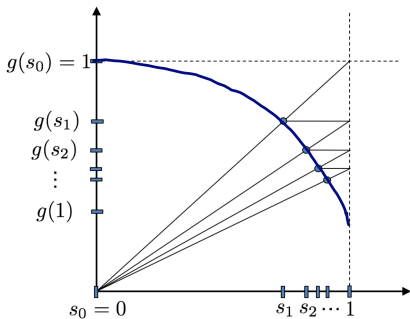
Under these assumptions it can be shown that the Hamiltonian

$$\tilde{H}(s) = \tilde{g}(s) |b\rangle \langle b| + (\mathbb{1} - A(s))$$

can be used instead with a total runtime  $T^* \leq 16\pi(\kappa + 1)\kappa^2/\epsilon$ .



# Construction of $g(s)$



$g(s)$  is monotonically decreasing.  
Given  $s_{k-1}$ ,  $\exists s_k > s_{k-1}$  such that

$$g(s_{k-1}) s_k = g(s_k)$$

$s_0 = 0 < s_1 < \dots < s_K < \dots < 1$   
converges to 1, s.t  $K = O(\kappa \log(\kappa/\epsilon))$   
steps are enough.

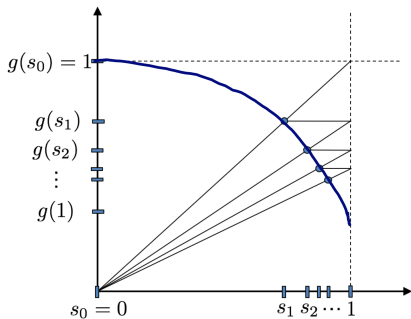
Define sequence of Hamiltonians

$$H_k := g(s_{k-1}) |b\rangle\langle b| + (\mathbb{1} - A) = \frac{g(s_{k-1})}{g(s_k)} [g(s_k) |b\rangle\langle b| + s_k (\mathbb{1} - A)] = \frac{g(s_{k-1})}{g(s_k)} H(s_k)$$

Has the same eigenvectors as  $H(s_k)$ .

Eigenvalue of  $|x(s_k)\rangle$  is given by  $g(s_{k-1})/g(s_k) = s_k^{-1}$ .





$$g(s_{k-1}) s_k = g(s_k)$$

$$H_k := g(s_{k-1}) |b\rangle \langle b| + (\mathbb{1} - A)$$

Assume we know  $s_{k-1}$  and  $g(s_{k-1})$ .

We first prepare  $|x(s_k)\rangle$  using  $H_k$ .

We then measure the expectation value  $H_k$  to obtain

$$s_k = \langle x(s_k) | H_k | x(s_k) \rangle^{-1}$$

From  $g(s_{k-1})$  and  $s_k$  we can learn

$$g(s_k) = g(s_{k-1}) s_k$$

$K = O(\kappa \log(\kappa/\epsilon))$  steps are enough.



# Adiabatic Quantum Computation

The evolution induced by the RM is closely related to the coherent evolution induced by the quantum adiabatic method.

assuming  $g(s)$  is known ;  $T^* \sim \kappa \log(\kappa)/\epsilon$

$$s_j = \frac{1 - (1/\kappa)^{j/q}}{1 - 1/\kappa} \longrightarrow s(t) = \frac{1 - (1/\kappa)^{t/T^*}}{1 - 1/\kappa}$$

assuming  $g(1)$  is known ;  $T^* \sim \kappa^3/\epsilon$

$$s_j = \frac{\kappa j/q}{1 + (\kappa - 1)j/q} \longrightarrow s(t) = \frac{\kappa t/T^*}{1 + (\kappa - 1)t/T^*}$$

assuming nothing ;  $O(\kappa \log(\kappa/\epsilon))$  iterations.

While quantum adiabatic approximations that imply these schedules are unknown, they are suggested by the Randomization Method.



## Summary

- A certain class of well-conditioned linear system of equations can be solved exponentially faster using quantum computers.
- The HHL algorithm and an algorithm due to Childs, Kothari and Somma runs on universal gate-based quantum computers.
- The adiabatic algorithms require evolutions with (and measurement of expectation values of) Hamiltonians that are linear combinations of  $A$  and  $|b\rangle\langle b|$ .
- The adiabatic algorithm is important in that this problem could be solved using a restricted, maybe non-universal quantum computing device.

